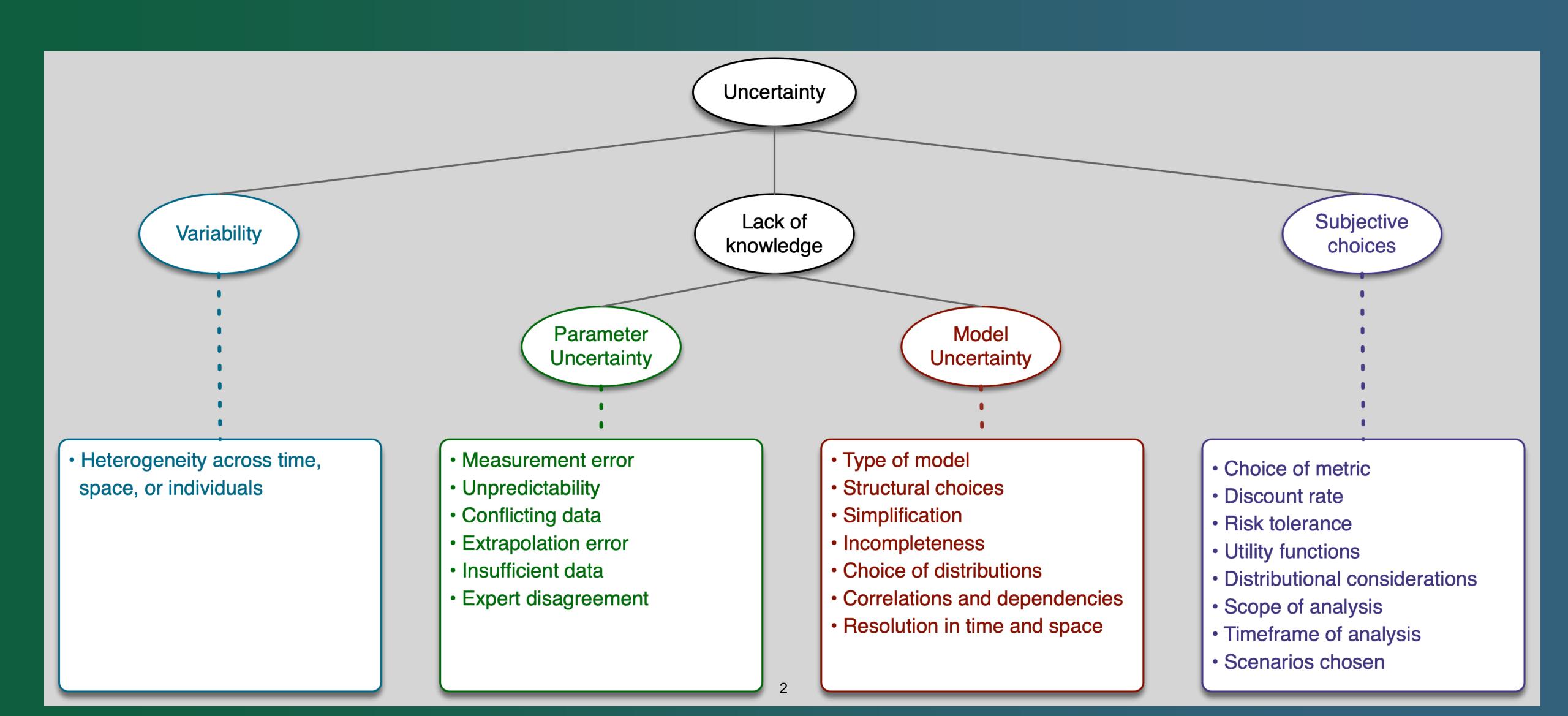
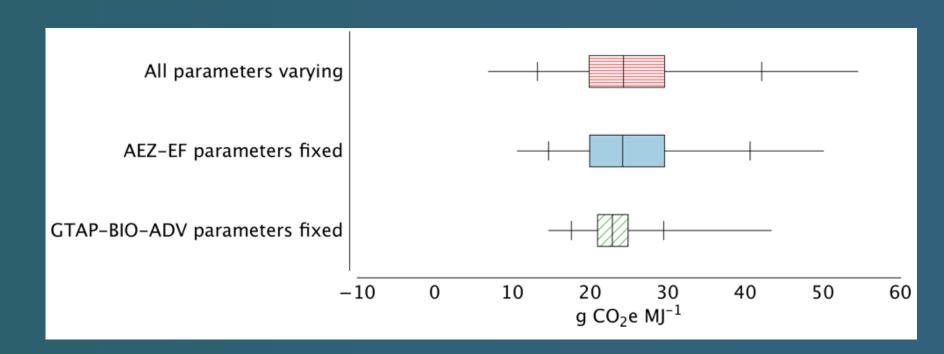


# What do we mean by uncertainty?

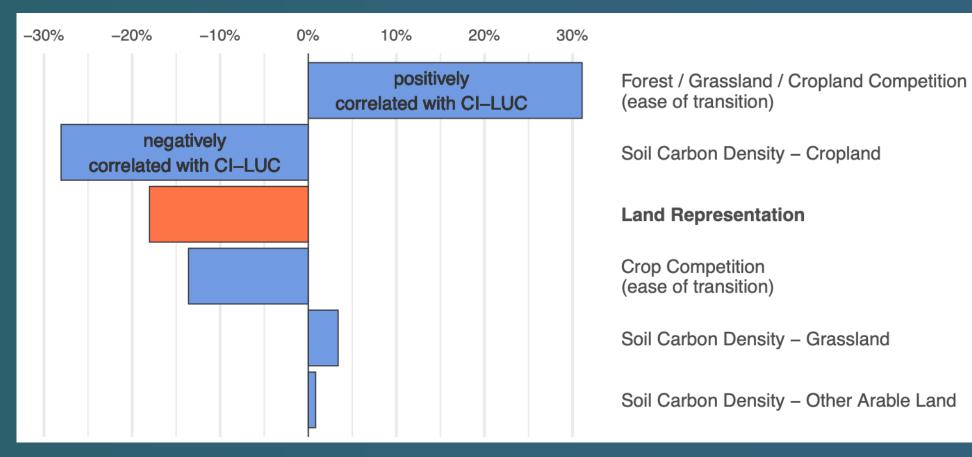


# Uncertainty & sensitivity analysis

- Uncertainty quantification
  - Distribution of model results across alternative parameters & choices
  - Statistics describing distribution of model results
- Global sensitivity analysis (SA)
  - Shows relative influence of parameters & choices on model results
  - Global SA accounts for parameter interactions across their ranges
    - One-at-a-time SA fails to account for these
- Ensemble analyses (e.g., Monte Carlo simulation) can inform both of these



Plevin, R.J., et al. 2015. 'Carbon accounting and economic model uncertainty of emissions from biofuels-induced land use change', Environmental Science & Technology, 49: 2656–64.



Plevin, R.J., et al. In review. 'Choices in land representation materially affect modeled biofuel carbon intensity estimates'.

# What uncertainty analysis doesn't do

- Global economic / ecosystem models are not truth machines
  - Many simplifications, deletions, and distortions
  - Can't predict non-stationary, complex, open systems

Oreskes, N. et al., 1994. 'Verification, Validation, and Confirmation of Numerical Models in the Earth Sciences', Science, 263: 641-46.

#### Therefore,

- Output distribution is not a (real world) probability distribution
  - It describes behavior of the model as constructed
- A range of model results may not bound real world outcomes

# Match methods to purpose

- What is the goal of our analysis and how will the result be used?
  - Choices: methods, scenarios, scope, resolution, timeframe, etc.
- Are we trying to:
  - 1. Estimate climate change mitigation from biofuel programs?
  - 2. Produce a CI value for use in a regulation?
- Models designed for (2) generally do not answer (1)
  - Different purposes require different analyses

Plevin, R.J., 2014. 'Using Attributional Life Cycle Assessment to Estimate Climate-Change Mitigation Benefits Misleads Policy Makers', Journal of Industrial Ecology, 18: 73-83.

# Mitigation vs carbon intensity

To estimate climate change mitigation:

- Characterize effects on climate of an action compared to BAU
- Be comprehensive to avoid unintended consequences
- Improve model whenever data or scientific understanding allow

When estimating regulatory carbon intensity values:

- Methods may be prescribed in legislation
- Model updates causing large changes in results are politically fraught
- Avoid claims about mitigation not supported by this analysis

# What is carbon intensity?

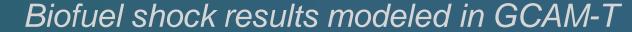
- CI has no well-established, concrete definition
- Every regulatory model of CI defines it differently
  - Different methods, models, boundaries, assumptions, data, timeframes

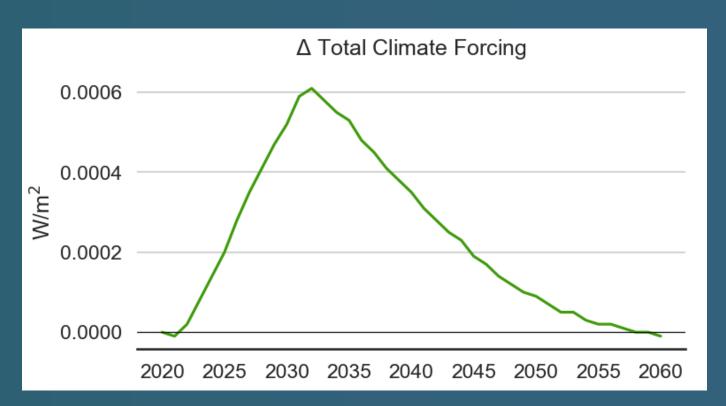
#### Implications:

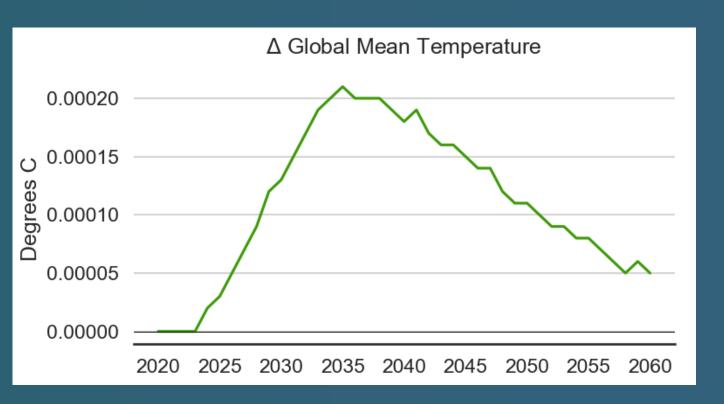
- Cl is scenario & model dependent; it's not a concrete fuel property
- Range of CI results reflects disagreement more than uncertainty
- Results using different definitions of CI are incommensurable

# Estimating climate change mitigation

- Include known climate forcings and their uncertainties
  - GHGs
  - GHG-precursors (e.g., CO, VOC)
  - aerosols (e.g., black carbon, organic carbon, SO<sub>X</sub>)
  - albedo change (e.g., resulting from LUC)
- CO<sub>2</sub>-equivalence of regional effects is not straightforward
- Better to aggregate as radiative forcing or temperature?
- Ignoring uncertain factors doesn't reduce uncertainty; it hides it







# Many subjective decisions are required

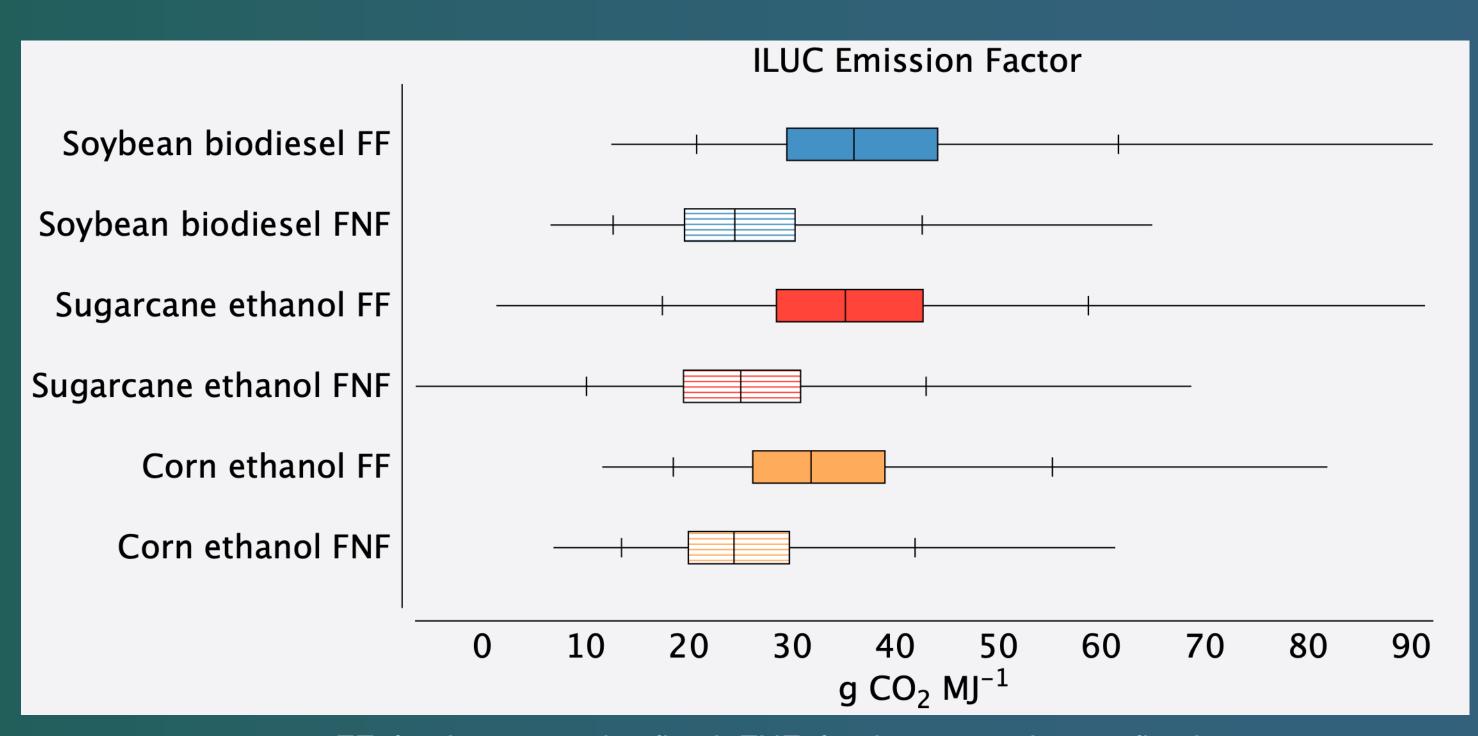
- Baseline scenario
- Analytic horizon or date of reckoning
- Size and shape of biofuel shock
- Climate effects to include
- Climate effects aggregation method (GWP, GTP, ΔRF, ΔΤ)
- Type of model (dynamic or static, myopic or foresight, partial or general equilibrium)
- Model resolution (sectors, regions, land types, technologies, time step)
- Focus of analysis (product vs policy)

### Best practices for modelers

- Sensitivity analysis is one of the "legitimate uses of a model" (Saltelli, et al. 2000)
  - Use global SA to capture parameter interactions
  - Use SA to interrogate your model, not "prove" it robust (Saltelli, 2010)
- Identify uncertainties strongly influencing variability in model results
- Demonstrate affects of subjective model choices on model results
- Avoid characterizing model results as predictions about real world
- Avoid unwarranted precision when presenting model results
- Document model limitations, assumptions, unquantified uncertainties

## Example 1

- ILUC analysis with GTAP-BIO and AEZ-EF models
- Monte Carlo simulation
  - 3 biofuels
  - 2 model structures (food consumption constraint)
- Results presented as distributions
- Model limitations explained

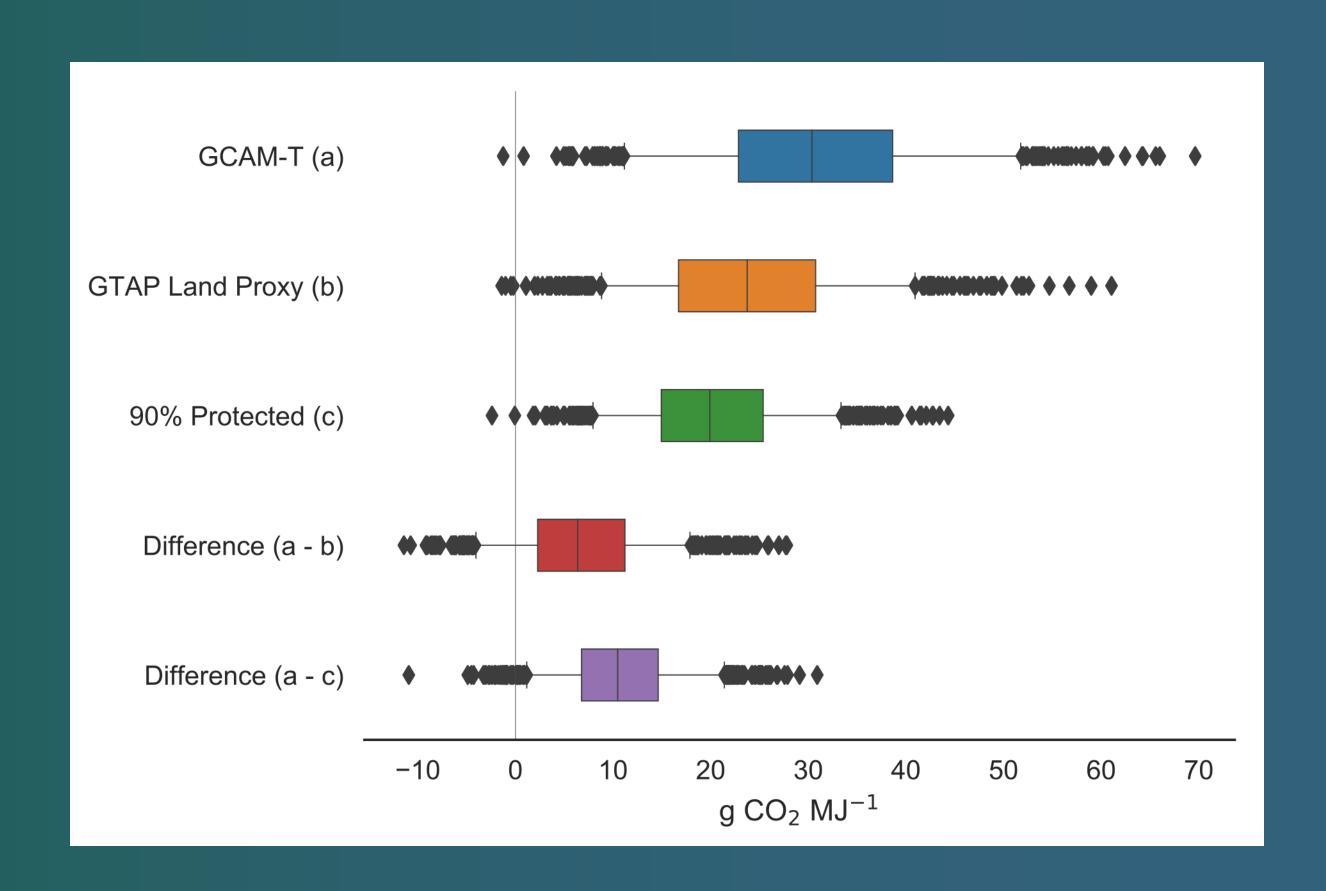


FF: food consumption fixed; FNF: food consumption not fixed

Plevin, R.J., et al. 2015. 'Carbon accounting and economic model uncertainty of emissions from biofuels-induced land use change', Environmental Science & Technology, 49: 2656–64.

# Example 2

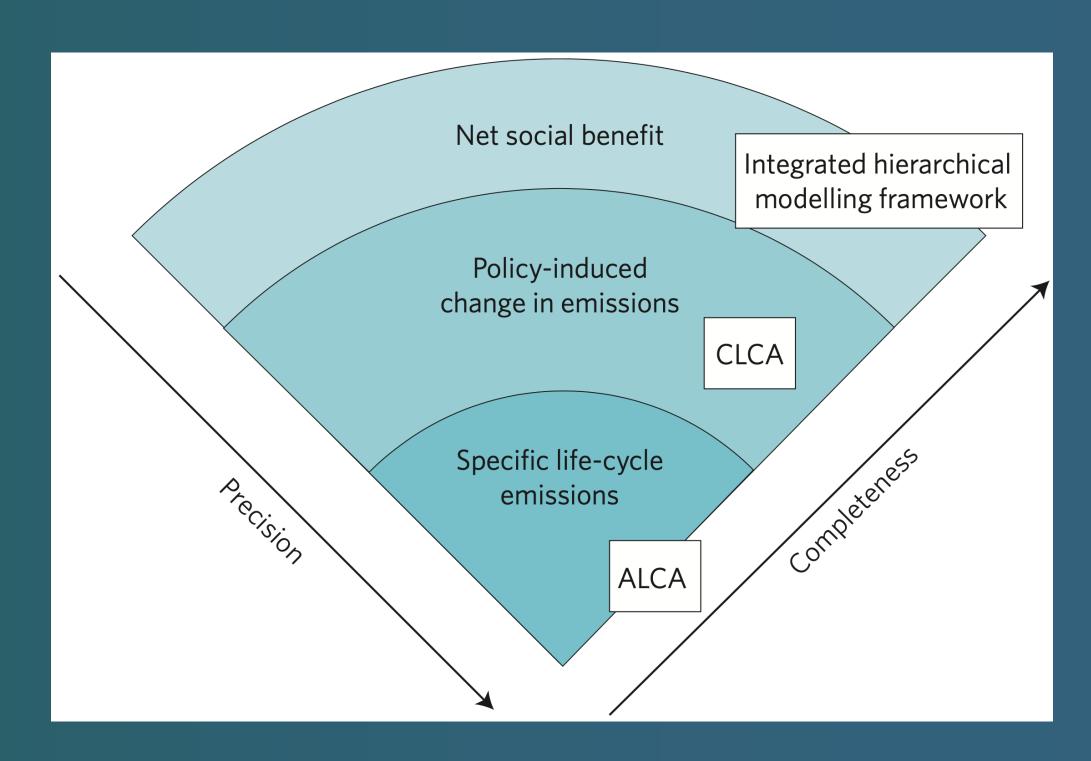
- Effect of land representation on LUC Cl using GCAM
  - 3 different land representations
  - Monte Carlo simulation
- Presents distributions per model and for per-trial differences
- Avoids claims about real world outcomes



Plevin, R.J., et al. In review. 'Choices in land representation materially affect modeled biofuel carbon intensity estimates'.

# Be forthright about model limitations

- The model is not the real world
- Subjective choices often drive results
- ILUC isn't the only market-mediated effect
- Actual net petroleum displacement is key determinant of mitigation (rebound effect)
- Cannot compare effects of biofuel with fossil fuel
  CI; oil displacement is one of these effects
- Models designed for one purpose may have blind spots when used for another purpose
- Excluding uncertain features doesn't reduce uncertainty
- Uncertainty increases with scope of model



Creutzig, F., et al. 2012. 'Reconciling top-down and bottom-up modeling on future bioenergy deployment', Nature Clim. Change, 2: 320-27.